

# Quick Review on Linear Space

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## 1 Linear Space Definition

The set  $V$  is a linear space if

1. Addition is commutative :  $\forall u, v \in V, u + v = v + u$
2. Addition is associative :  $\forall u, v, w \in V, (u + v) + w = u + (v + w) = v + (u + w)$
3. Identity element in addition :  $\exists! O \in V$  s.t.  $\forall u \in V, u + O = u$
4. Negative :  $\exists! v \in V$  s.t.  $\forall u \in V, u + v = O$
5. Scalar multiplication is commutative, associative, distributive :  $\forall a, b \in \mathbb{R}$  and  $\forall u, v \in V, a(bu) = (ab)u = b(au), (a + b)u = au + bu, a(u + v) = au + av$
6. Identity element in scalar multiplication :  $\exists! 1 \in \mathbb{R}$  s.t.  $\forall u \in V, 1u = u$

## 2 Subspace

For nonempty subset  $S$  of linear space  $V$ , if  $S$  is

1. Close in addition :  $\forall u, v \in S$ , then  $u + v \in S$
2. Close in scalar multiplication :  $\forall a, u \in S$ , then  $au \in S$

Then  $S$  is a subspace of  $V$

## 3 Linear Combination, Linear Independence & Dependence

For vector  $u, v$ , their linear combination  $w$ , for any value of  $a$  and  $b$  is  $w = av + bu$

For a linear space  $V$ , a set of vector  $v_1, v_2, \dots, v_n$  is linear dependence if there exists a group of non-zero element  $a_1, a_2, \dots, a_n$  in  $\mathbb{R}$  that

$$\sum_{i=1}^n a_i v_i = O$$

If there is no such set of  $\{a_i\}$ , then  $\{v_i\}$  are all linear independent

Geometrically, it means if a vector can not be expressed as a linear combination of other vectors, then such vector is linear independent to other vector, and if such vector can be expressed as a linear combination of other vectors, then such vector is linear dependent to other vector.

## 4 Span, Basis, Dimension

A vector  $w$  is span by  $u, v$  if  $w$  can be expressed as a linear combination of  $u$  and  $v$

Geometrically, that means  $w$  can be represented / generated / formed by  $au + bv$  for some  $a, b$

For a set of vectors  $\{v_i\}_{i=1}^{i=n}$  in a linear space  $V$ , if

1.  $v_1, v_2, \dots, v_n$  are all linearly independent
2.  $v_1, v_2, \dots, v_n$  span  $V : \text{span}\{v_1, v_2, \dots, v_n\} = V$

Then  $\{v_i\}_{i=1}^{i=n}$  is a basis for  $V$

For linear space  $V$ , if  $V$  has a basis of  $n$  vectors, then dimension of  $V$  is  $n$

## 5 Kernel / Null space

The kernel, or the null space of a matrix  $A$  is a set that it contains all the element  $x$  s.t.  $Ax = 0$

That is,  $\text{Ker}(A) = \{x \in V : Ax = 0\}$

## 6 Rank

For a matrix  $A$ ,

Column rank( $A$ ) = Row rank( $A$ ) = Dimension of largest non-singular submatrix of  $A$   
= Number of pivot of  $A$

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